

# Data compression algorithms for flow tables in Network Processor RuNPU.

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**Abstract**—This paper addresses the problem of packet classification within a network processor (NP) architecture without the separate associative device. By the classification, we mean the process of identifying a packet by its header. The classification stage requires the implementation of data structures to store the flow tables. In our work we consider the NP without the associative memory. Flow tables are represented by an assembly language program in the NP. For translating flow tables into assembly language programs, a tables translator was used. The main reason for implementing data compression algorithms in a flow tables translator is that modern flow tables can take up to tens of megabytes. In this paper we describe the following data compression algorithms: Optimal rule caching, recursive end-point cutting and common data compression algorithms. An evaluation of the implemented data compression algorithms was performed on a simulation model of the NP.

**Index Terms**—Network processor, software-defined networks, packet classification, data compression.

## I. INTRODUCTION

At present, software-defined networks (SDN) are in active development and require high-performance switches [5]. The main functional element of the high-performance SDN switch is a programmable network processor (NP). The network processor is a system-on-chip specialized for network packet processing. In this work we consider a programmable NP. A programmable NP is one that supports on-the-fly modification of the packet processing program and the set of header fields to be processed.

In this article we discuss data compression algorithms used for flow tables. Flow tables are needed for packet classification process. A flow table is the set of rules defined by OpenFlow protocol. OpenFlow is one of the most common protocols for controlling a network SDN switch. This paper considers OpenFlow version 1.3 [3]. Each rule contains a match field, a bit string by which a packet can be identified and a set of actions, that the NP performs on this packet. Classification is the process of the identification of a network packet by its header.

This article has the following structure: in second section we introduce problem, in third section we introduce the NP architecture and flow tables translator, in fourth section we describe related work, in fifth section we

describe data compression algorithm implementation and in sixth section we introduce our evaluation methodology.

## II. THE PROBLEM

Let us consider OpenFlow tables formalisation. An ordered set of all considered attributes is denoted as  $I = \{m_1, m_2, \dots, m_k\}$ . Every attribute  $m_i$  from the set  $I$  is described by a bit string  $m_i \in \{0, 1, *\}_i^W$ . In this article symbol  $*$  denotes any bit. But, if  $\exists m_i^j \in m_i$  and  $m_i^j = *$ , then for  $\forall m_i^k$ , where  $k > j$ ,  $m_i^k = *$ . The length of the attribute is denoted  $len(m_i) = W_i$ .

Flow tables are represented by a set of rules  $R = \{r_1, r_2, \dots, r_n\}$ . With every rule  $r_i$  binding the features:

- An index  $i$ ;
- A priority  $p_i \in \mathbb{Z}_+$ ;
- A vector of values of attributes  $f_i = \{f_i^1, f_i^2, \dots, f_i^k\}$ , where  $f_i^j$  is an attribute value  $m_j \in I$ .
- A set of actions  $A_i = \{a_1, a_2, \dots, a_z\}$ .

A network packet header  $x$  and its metadata with vector values of attributes  $g = \{g^1, g^2, \dots, g^k\}$  ( $x \rightarrow g$ ), a match rule  $r_i \in R$  with a vector of values of the attributes  $f_i = \{f_i^1, f_i^2, \dots, f_i^k\}$  and a priority  $p_i$  (a rule  $r_i \in R$  identifies a network packet with a vector values of attributes  $g$ ), if:

- 1) a vector values of attributes  $g$  match a vector of values of the attributes  $f_i$ ,  $\forall g_i \in g, len(g_i) = len(f_i)$ .  $\forall f_i^{lj} \in f_i^l, f_i^{lj} \in \{*, g^{lj}\}, l = \overline{1, k}$ ;
- 2) a priority  $p_i$  is the highest among all rules  $r_j \in R$ , if a vector  $g$  match a vector  $f_j$ .

The set of rules  $R$  must satisfy the following constraint. For any two rules  $r_i, r_j \in R, r_i \neq r_j$ , if their vectors of values intersect, there is a set of attribute values. This set corresponds to vectors of values of attributes of both rules  $p_i \neq p_j$ .

Let us introduce the function for network packet identification  $x \rightarrow g$  in flow table  $R$ , (denotes as  $R(x)$ ). It returns a set of actions, that corresponded to the rule  $x \rightarrow g$ .

$R(x) = A_{r_i}$ , where  $A_{r_i}$  is the set of actions  $r_i \in R$ .

We need to introduce a **similar concept** of the sets of rules  $R_1$  and  $R_2$ . The set  $R_1$  is similar to the set  $R_2$ , when for any network packet header, that can be identified by some rule from the set  $r_i \in R_1$ , and there exists another rule that identifies it as  $r_j \in R_2$ , and  $A_i = A_j$ .

We need to develop an algorithm for compressing flow tables. This algorithm must translate an input flow table (a set of rules  $R_1$ ) into a new compressed set of rules  $R_2$ .

- 1) The set of rules  $R_1$  is similar to the set of rules  $R_2$ .
- 2) The cardinality of the set  $R_2$  must be lower than the cardinality of the set  $R_1$ .

### III. NETWORK PROCESSOR ARCHITECTURE

In the considered NP the pipeline architecture is used, with each pipeline consisting of 10 computing blocks. To avoid complex memory organization, there is no associative memory in the considered NP. The NP uses the same memory both for commands and data.

Let us consider the pipeline NP architecture. Each computing block has an access to the memory area where the program with data is located. There is a limit of 25 clock cycles per packet on each processing block. There is up to 512 kilobytes to store assembly language program representing flow tables. Due to the instruction set architecture, there is no separate memory area where data is stored. Therefore, the microcode contains all the data, required to classify packets.

#### A. Flow tables translator

Flow tables translator is a tool that is executed on CPU. It is used for flow tables translating into assembly language programs, that can be interpreted by NP. Flow tables translator uses tree structures for flow table representation. Every node of the tree structure can be associated with a table rule. After building a tree every node is translated into a part of an assembly language program. Here is a flow tables translator workflow:

- 1) Load a flow table from file.
- 2) Check every rule in the table.
- 3) Build a tree structure from a set of rules.
- 4) Translate every node into a part of an assembly language program.
- 5) Combine all translated parts into the one assembly language program.
- 6) Add a header that corresponds to used protocol.
- 7) Write the assembly language program into file.

This tool was implemented in work [4].

### IV. RELATED WORK

In this section we introduce a review on data compression algorithms, that already used for other network processors [1]. To choose algorithms for implementation in NP we used the following criteria:

- 1) Compression rate, is needed for algorithm performance evaluation.
- 2) Evaluation of compression algorithm complexity.
- 3) Usability of compressed flow tables without decompression.
- 4) The necessity to use external memory by the algorithm.

#### A. Most common data compression algorithms

Data compression algorithms have evolved over the years. Nowadays compression algorithms can be used in many different ways. In this section we describe the algorithms that compress data in binary format. There are most known of them: Huffman coding, JPEG, LWZ, zip. These algorithms require decompression for data usage. And this is why we will not use them in our flow table translator.

#### B. Optimal rule caching

Optimal rule caching algorithm is more specific data compression algorithm. It is used for table compressing in SND switches [6]. It is based on search tree structure, that is built based on rules usage frequencies. There are two trees: the first tree consist from most used rules. This tree is translated into assembly language program. The second tree consist from other rules, it is stored in CPU memory.

#### C. Recursive end point cutting

Recursive end-point cutting algorithm is based on HyperSplit tree usage. Compressing is performed by destroying duplication rules [2]. This algorithm permits operations with flow tables without full rebuilding tree. By rules duplication we understand the following rules:

- A rule, storing in node duplicates of the rules in leaf nodes. (particle duplication).
- A rule, storing in node duplicates of the rules in all leafs nodes. (full duplicating rule).

This algorithm recursively uses NewHypersplit tree to remove duplicate rules from the currently being built tree. The deleted duplicate rules are then collected into a second rule table, called a recursive table, to build a second tree. It is possible that duplicate rules still exist in the second tree, and some of them are also removed and used to build the third tree.

This tree building process is performed recursively while there are duplicate rules in the last tree.

### V. ALGORITHMS COMPARISON

Let us describe data compression algorithms comparison in table I. Each algorithm has its own pros and cons.

- 1) **Optimal rule caching** — has the highest compression ratio, and is quickly implemented in the considered NP. The need to use external memory imposes additional overhead on some packet processing.
- 2) **Recursive end-point cutting** — has the lowest compression ratio, it is more difficult to implement than the optimal rule caching algorithm. Moreover, this algorithm does not require the use of external memory.
- 3) **Common data compression algorithms** — have good compression ratios on average, but require data decompression.

TABLE I: Data compression algorithms comparison

Name	complexity of construction	Compression rate	CPU memory usage	Decompression needed
Optimal rule caching	$O(N^2)$	0.1...0.9	yes	no
Recursive end-point cutting	$O(N * \log(N))$	0.1	no	no
Common data compression algorithms	$O(K * \log_2 N)$	0.1...0.8	no	yes

## VI. OUR SOLUTION

In this section we introduce our solution of flow tables compressing.

### A. Flow table optimization

First of all we need to introduce operation **getting last important bit**  $last(m_i) = j$ ,  $m_i^j \in \{0, 1\}$  and  $m_i^{(j+1)} = *$ . We claim that the rules  $r_i \in R$  and  $r_j \in R$  are the **same**, if  $\forall u \in len(f_i) \ last(f_i^u) = last(f_j^u) = l$ , but  $f_i^{ul} \neq f_j^{ul}$  and  $A_i = A_j$ . For flow table optimization we need to remove all **same** rules.

### B. Main flow table compression algorithm

Let us introduce a packet header distribution  $P$ , where  $p_x$  mean network packet income probability  $x \rightarrow g = \{g^1, g^2, \dots, g^k\}$ . We need a correction ratio  $T_P(R_1, R_2)$ , where  $R_1$  and  $R_2$  are two different flow tables. Thus correction ratio means probability of incoming network packet header by distribution  $P$ . As well as probability of identifying this network packet by rules  $r_1 \in R_1$  and  $r_2 \in R_2$ . Moreover the sets of actions of this rules are similar  $A_1 = A_2$ ,  $A_1 \in r_1$ ,  $A_2 \in r_2$ .

$$T_P(R_1, R_2) = \sum_{x \rightarrow g, R_1(x)=R_2(x)} p_x$$

The optimal correction ratio for flow table  $R$  and a number of rules  $n$  and a network packet header distribution  $P$  is:

$$\zeta(n, R, P) = \max_{R_i, |R_i| < n} T_P(R, R_i)$$

Let  $p^i$  be probability of choosing rule  $r_i \in R$ , in distribution  $P$ . Let rules in flow table  $R$  be in not-increasing order of their probabilities. Then:

$$\zeta(n, R, P) \geq \sum_{i \in [1, n]} p^i + 1 - \sum_{i \in [1, n_0]} p^i \geq n/n_0$$

This algorithm needs exploration and building a flow table  $R_a$ , based on input flow table  $R$ . There is a minimal set of rules ( $n_0$ ) and a maximum optimal correction ratio  $\zeta(n, R, P)$ .

### C. Software solution

In this section we introduce software workflow of our algorithm. First of all we need to add a new fields in tree structure nodes for our algorithm.

- A probability into tree node. It must be filled if node contains rule.

- A sum of probabilities of leaf nodes.

Let us introduce program operation for split tree.

- Generate a set of tree nodes.
- Sort this set in non-increasing order.
- Create a counter that stores a sum of node probabilities.
- Get the first node with maximum self probability.
- Increase the counter.
- Add this node into another set and remove from first.
- Repeat last three operations while counter less then 0.95.
- Build tree from second set of rules.

After performing this operations we get the set of nodes. We could build first tree from second set of rules and second tree from first set of nodes. After this we need to translate the first tree into an assembly language program.

1) *Notation used:* Let  $node1, node2$  — tree vertices,  $value$  — some feature value. Let's introduce the following notations:

- $Tree.root$  — the root node of the tree  $Tree$ .
- $node_1(value)$  — the descendant of the node  $node_1$ , connected to it by an arc with the mark value.
- $node_1.rules$  — set of rules corresponding to node  $node_1$ .
- $node_1.edges$  — set of marks of arcs coming from node  $node_1$ .
- $node_1.prob$  — an amount of probabilities of rules.
- $copy(node_1, val, node_2)$  — a procedure that adds to the node  $node_1$  a descendant with an arc marked value, copying the tree that forms the node  $node_2$ .
- $equals(node_1, node_2)$  — function that returns *true* if the trees formed by nodes  $node_1$  and  $node_2$  are the same, otherwise it returns *false*. The comparison takes into account the rule sets and arc labels associated with the nodes.
- $same(rule_1, rule_2)$  — function that returns *true* if rules are **same**.
- $isleaf(node_1)$  — function that returns *true* if  $node_1$  — tree leaf, otherwise it returns *false*.

2) *Flow table optimization algorithm:* Let us introduce procedure **Same** (Listing 1), it returns a set of rules that are a union of **same** rules in the sets of two nodes.

Let us introduce flow table optimization algorithm. It can be describe by procedure **Optimize** (Listing 2), where **node** — tree node. For optimizing flow table we need to perform this procedure to  $Tree.root$ .

```

1 procedure Same(node_1, node_2):
2   rules = {}
3   for all rule_1 in node_1.rules do
4     for all rule_2 in node_2.rules do
5       if same(rule_1, rule_2) then
6         rules += {rule_1 union rule_2}
7       endif
8   return rules

```

Listing 1: Procedure for obtaining a set of rules derived from the same rules

```

1 procedure Optimize(node):
2   if not isleaf(node) then
3     for all val_1 in node.edges do
4       for all val_2 in node.edges do
5         if val_1 not equal val_2 then
6           node.rules += Same(node(val_1), node(val_2))
7         endif
8       for all val in node.edges do
9         Optimize(node(val))
10    endif

```

Listing 2: Procedure for optimizing the tree

## VII. EVALUATION

### A. Evaluation methodology

In this section we talk about evaluation methodology. Flow table compression algorithms can be evaluated by an assembly language program evaluation, from the flow tables translator with implementing this flow table compression algorithm. We used the following parameters in our evaluation:

- Memory usage by assembly language program.
- Average time of network packet processing in NP cycles.
- Maximum time of network packet processing in NP cycles.

The described analysis requires to execute the following actions for each flow table.

- 1) Choose a flow table for this experiment.
- 2) Translate the flow table using flow tables translator implementation based on: LPM tree, AVL tree, Our flow table compression algorithm.
- 3) Execute simulation on NP simulation model.
- 4) Evaluate results.

### B. Evaluation data

Several variants of the flow tables should be used for the evaluation. In this section, we will introduce the flow table templates.

- The first pattern — a flow table rule pattern contains the values of three attributes: input port number, destination MAC address and source MAC address.
- The second pattern — a flow table rule pattern contains the values of two attributes: IPv4 destination address and IPv4 source address.

- The third pattern — a flow table contain five attributes: input port number, destination MAC address, VLAN ID, L3-level header ID (EtherType) and destination IPv4 address.

There is example of input data (Listing 3).

```

1 {SRC_MAC , DST_MAC , INSTR}
2 {SRC_MAC , DST_MAC , INSTR}
3 +-----+
4 | 1 | :12      | :10:1 | goto_table 1 |
5 +-----+
6 | 1 | :23:45   | :20   | goto_table 1 |
7 +-----+
8 | 1 | :0       | :1    | goto_table 1 |
9 +-----+

```

Listing 3: Example of input data for input of a flow table

## VIII. FUTURE WORK

In the future we will refine evaluation data. We expect less memory usage with our compression algorithm implemented into flow table translator. In the first experiments we conducted, we obtained results showing a significant reduction in the amount of memory usage using the data compression algorithm. After this we could research possibility of TCAM memory implementation and use this compression algorithms for it.

## IX. ACKNOWLEDGEMENTS

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